



## Conditional Probability

### Introduction to Data Science Algorithms

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SLIDES ADAPTED FROM DAVE BLEI AND LAUREN HANNAH

## Administrivia

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- Autograder
- Office Hours
- Phone

## Context

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- Data science is often worried about “if-then” questions
  - If my e-mail looks like this, is it spam?
  - If I buy this stock, will my portfolio improve?
- Since data science uses the language of probabilities, we need *conditional* probabilities (continuing probability intro)
- Also need to **combine** distributions

## Conditional Probabilities

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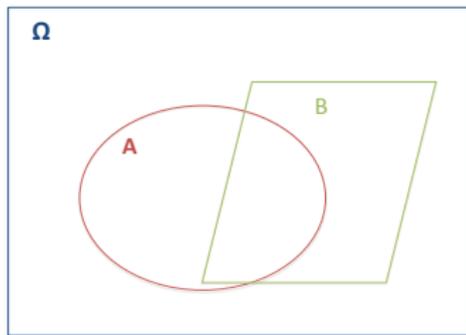
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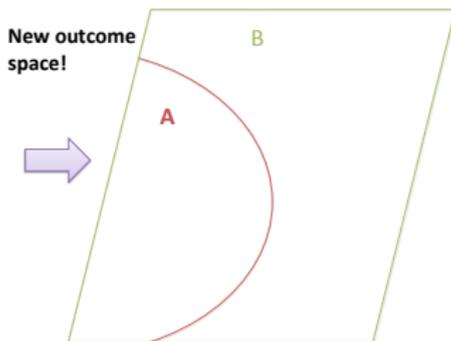
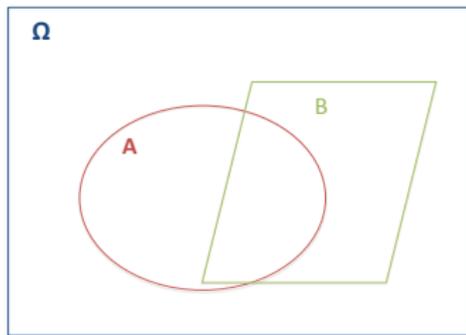


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## Independence (Reminder)

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Random variables  $X$  and  $Y$  are independent if and only if  $P(X = x, Y = y) = P(X = x)P(Y = y)$ . How does this interact with conditional probabilities?

Conditional probabilities equal unconditional probabilities with independence:

- $P(X = x | Y) = P(X = x)$
- *Knowing  $Y$  tells us nothing about  $X$*

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$$P(A > 3 \cap B + A = 6) =$$

$$P(A > 3) =$$

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## Combining Distributions

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- Sometimes distributions you have aren't what you need
  - Conditional  $\rightarrow$  joint (chain)
  - Reverse conditional direction (Bayes')

## The chain rule

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- The definition of conditional probability lets us derive the *chain rule*, which lets us define the joint distribution as a product of conditionals:

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- For example, let  $Y$  be a disease and  $X$  be a symptom. We may know  $P(X|Y)$  and  $P(Y)$  from data. Use the chain rule to obtain the probability of having the disease and the symptom.
- In general, for any set of  $N$  variables

$$P(X_1, \dots, X_N) = \prod_{n=1}^N P(X_n | X_1, \dots, X_{n-1})$$

## Bayes' Rule

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What is the relationship between  $P(A|B)$  and  $P(B|A)$ ?

$$P(B|A) = \frac{P(A|B)P(B)}{P(A)}$$

- 1 Start with  $P(A|B)$
- 2 Change outcome space from  $B$  to  $\Omega$
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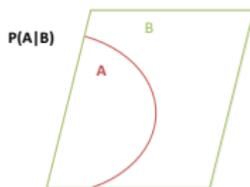
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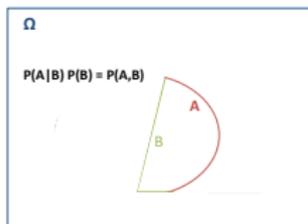
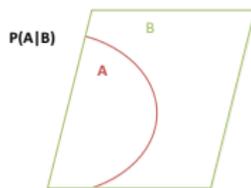
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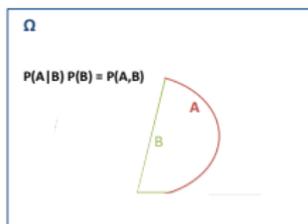
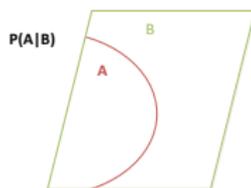
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$P(A|B)P(B)/P(A) = P(A,B)/P(A) = P(B|A)$

